

Modeling and Control of Manufacturing Systems

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From real fab to DEM

Effective process times

Control Framework

Approximation model

Control

Conclusions

Discrete Event Modeling of a real factory

- raw process time t_0 and c_0
- setups t_s and c_s
- TBF t_f and c_f , TTR t_r and c_r
- operator delays
- rework
- ...(!)

From real fab to DEM

Effective process times

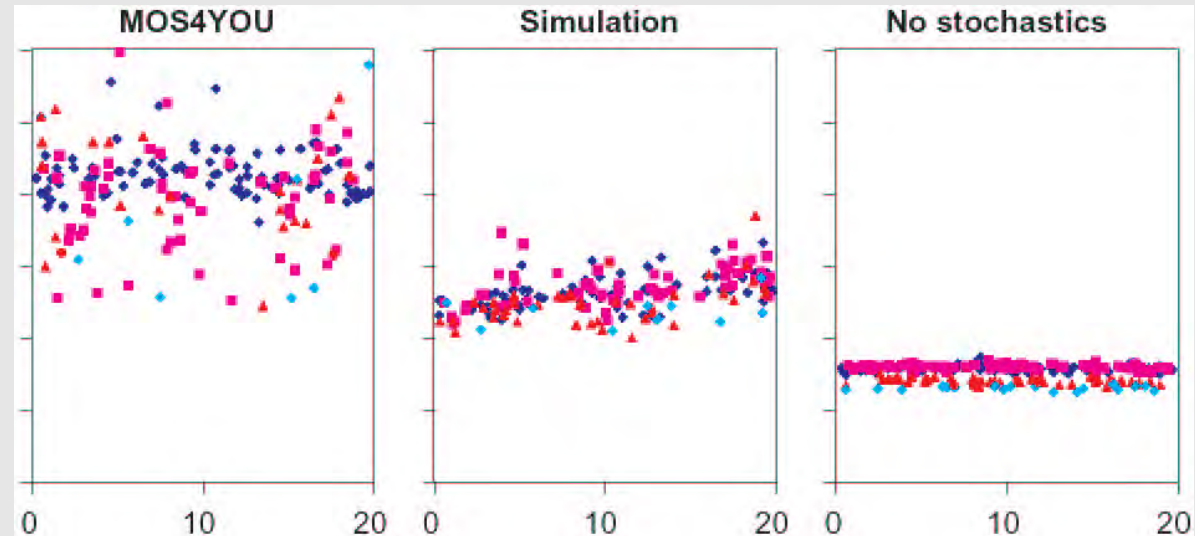
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Results



- smaller mean flow time
- smaller variance flow time

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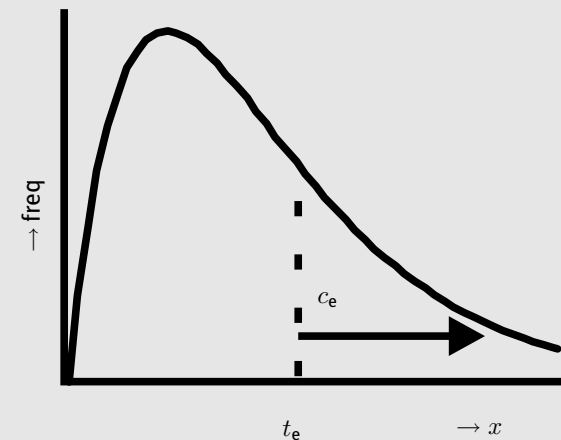
Conclusions

The effective process time method

- raw process time t_0 and c_0
- setups t_s and c_s
- TBF t_f and c_f , TTR t_r and c_r
- operator delays
- rework
- ... (!)

Idea:

Combine all disturbances in one single EPT probability density function



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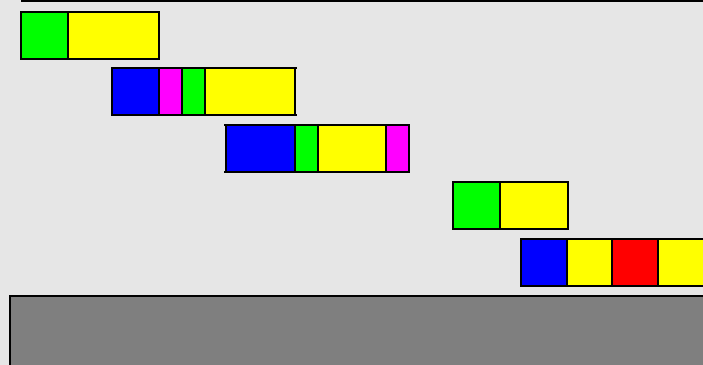
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0 5 10 15 20 25 30



Legend

Setup

Processing

Queueing

Waiting for operator

Machine breakdown

From real fab to DEM

Effective process times

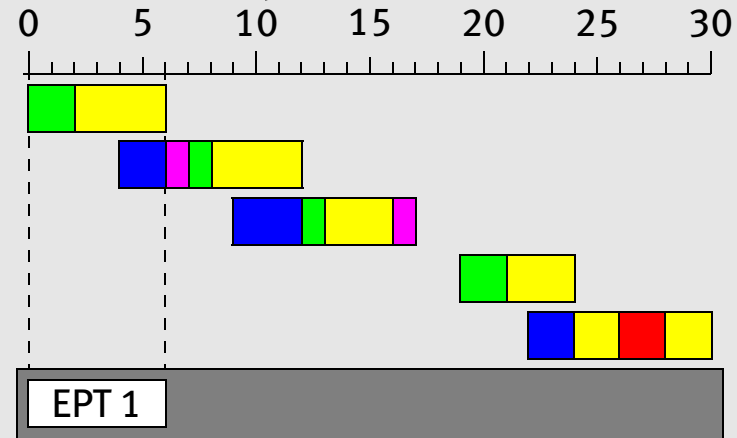
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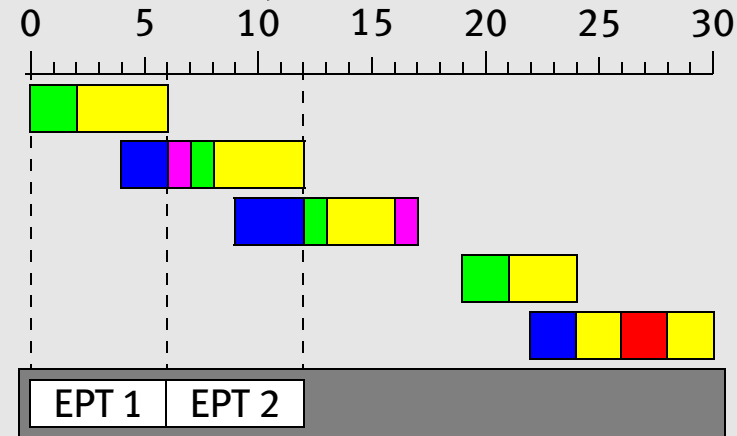
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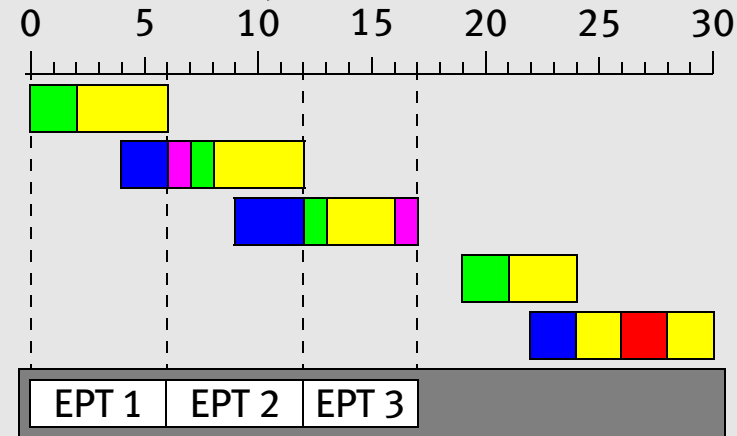
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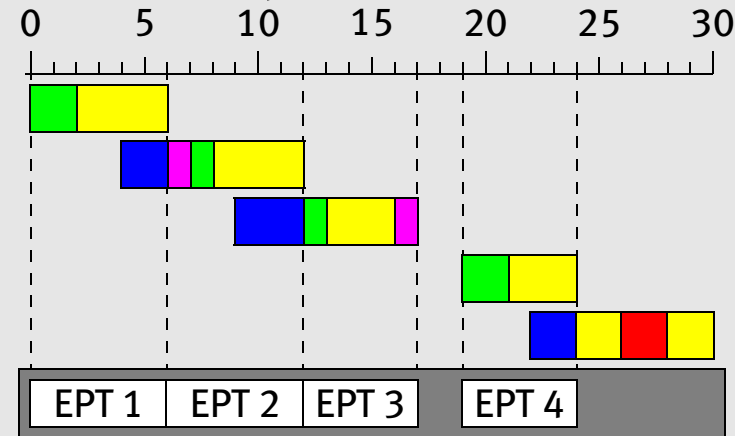
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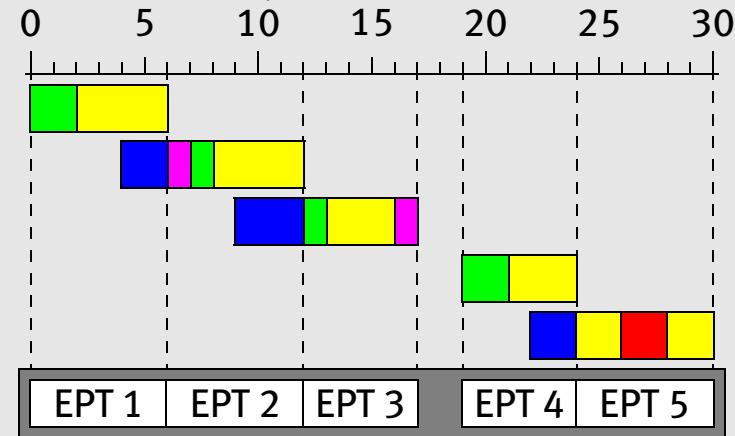


Legend

- Setup
- Processing
- Queueing
- Waiting for operator
- Machine breakdown

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Legend

- Setup
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- Waiting for operator
- Machine breakdown

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Manufacturing System

From real fab to DEM

Effective process times

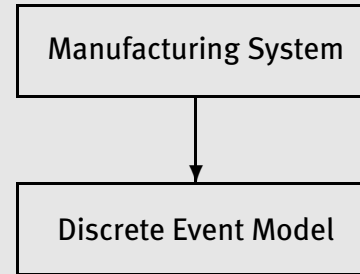
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From real fab to DEM

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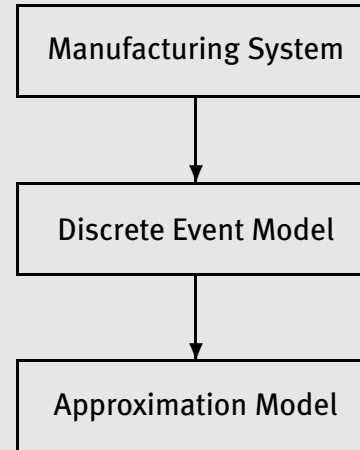
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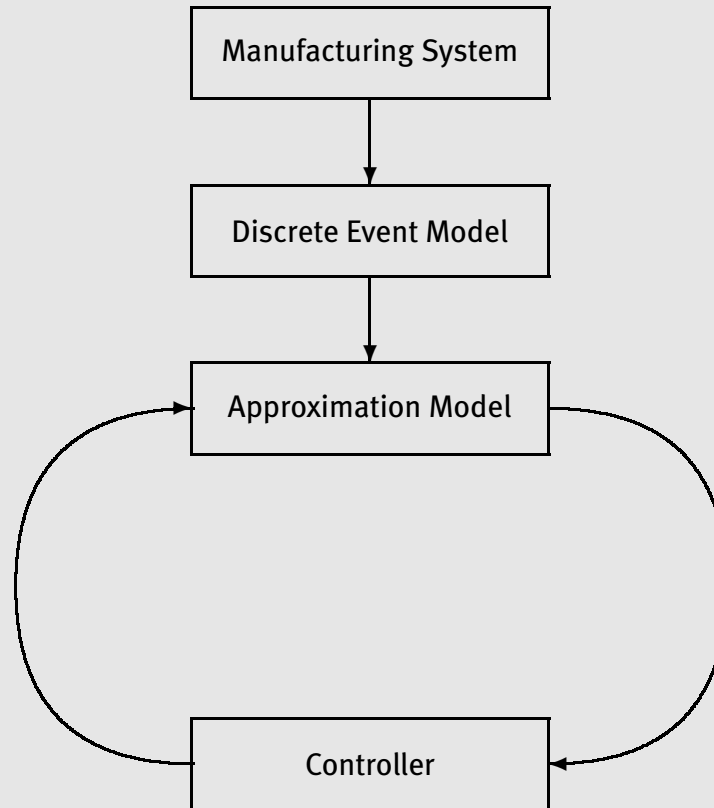
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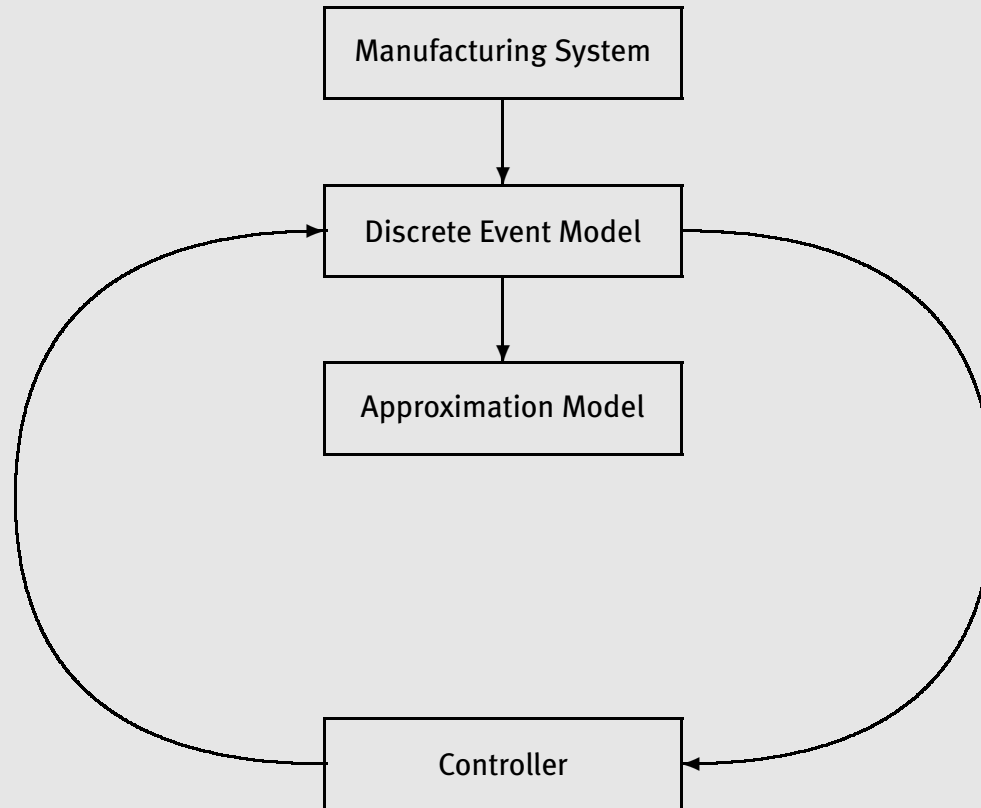
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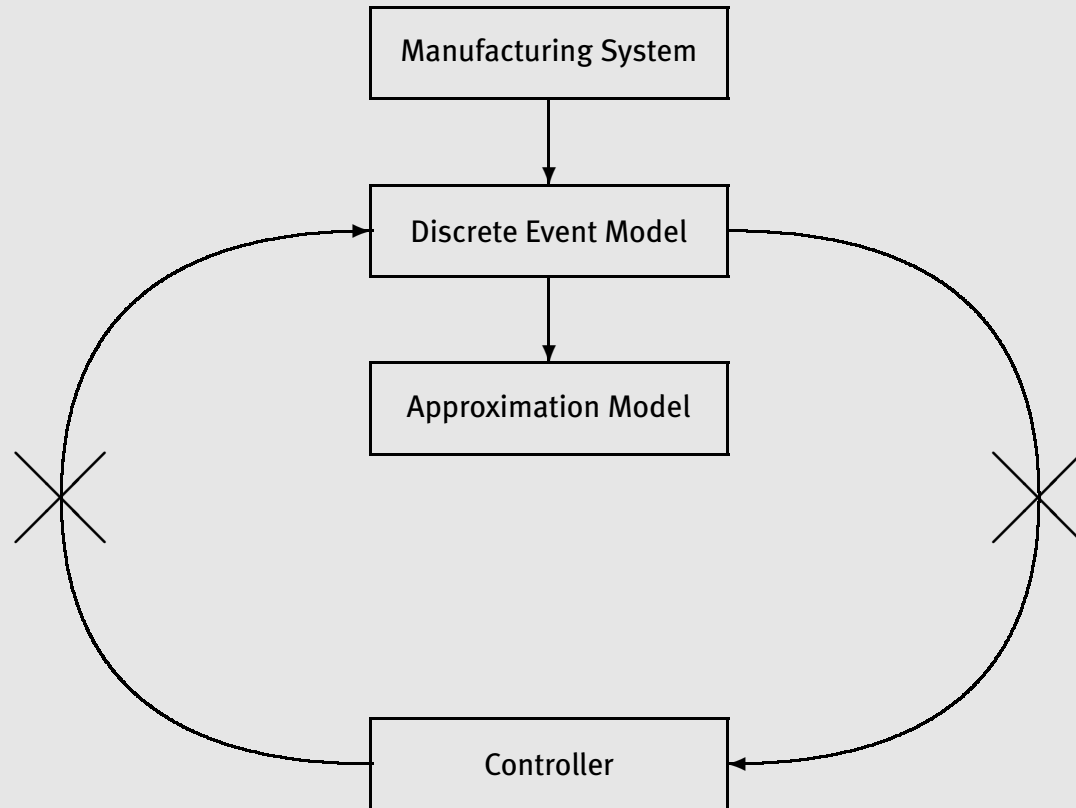
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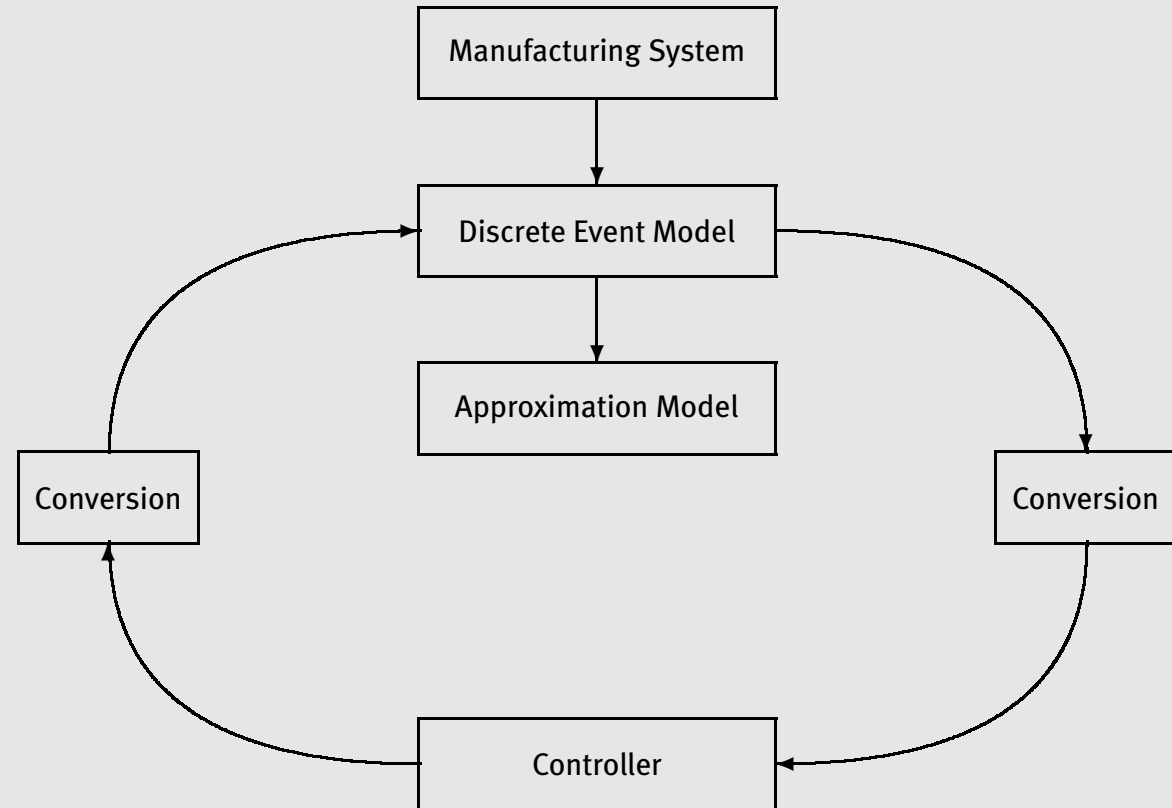
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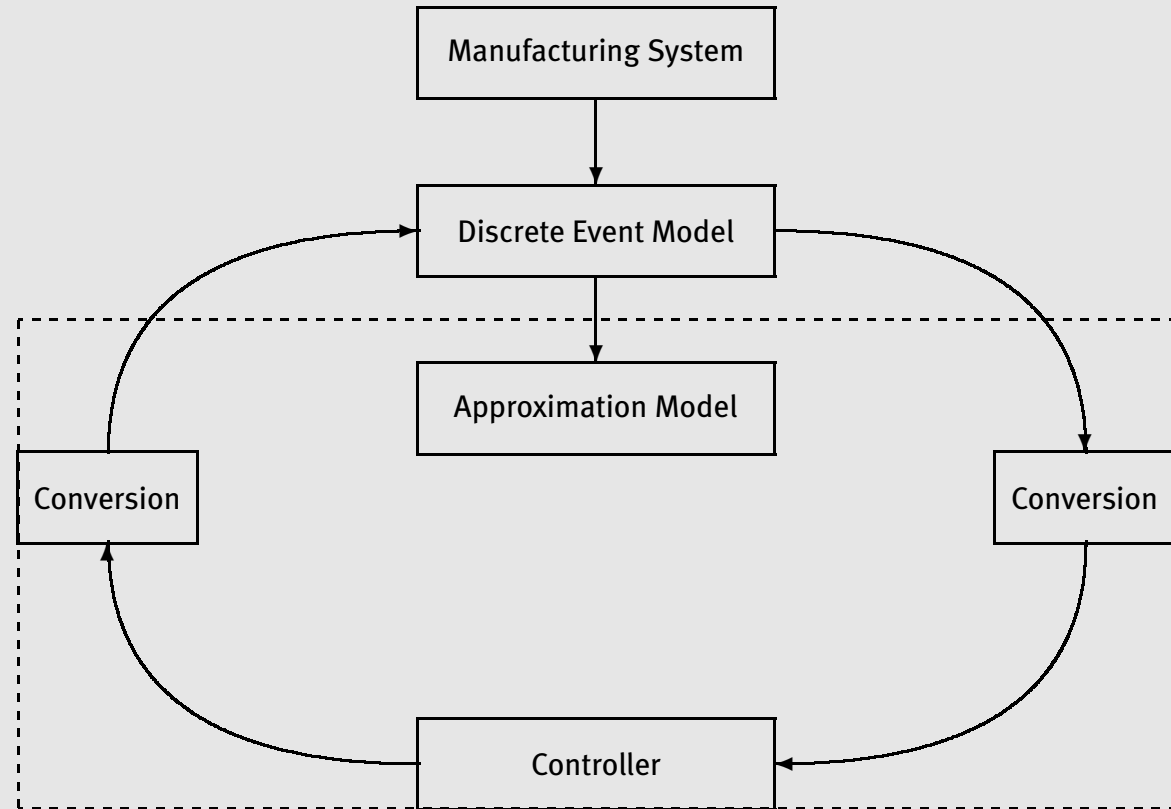
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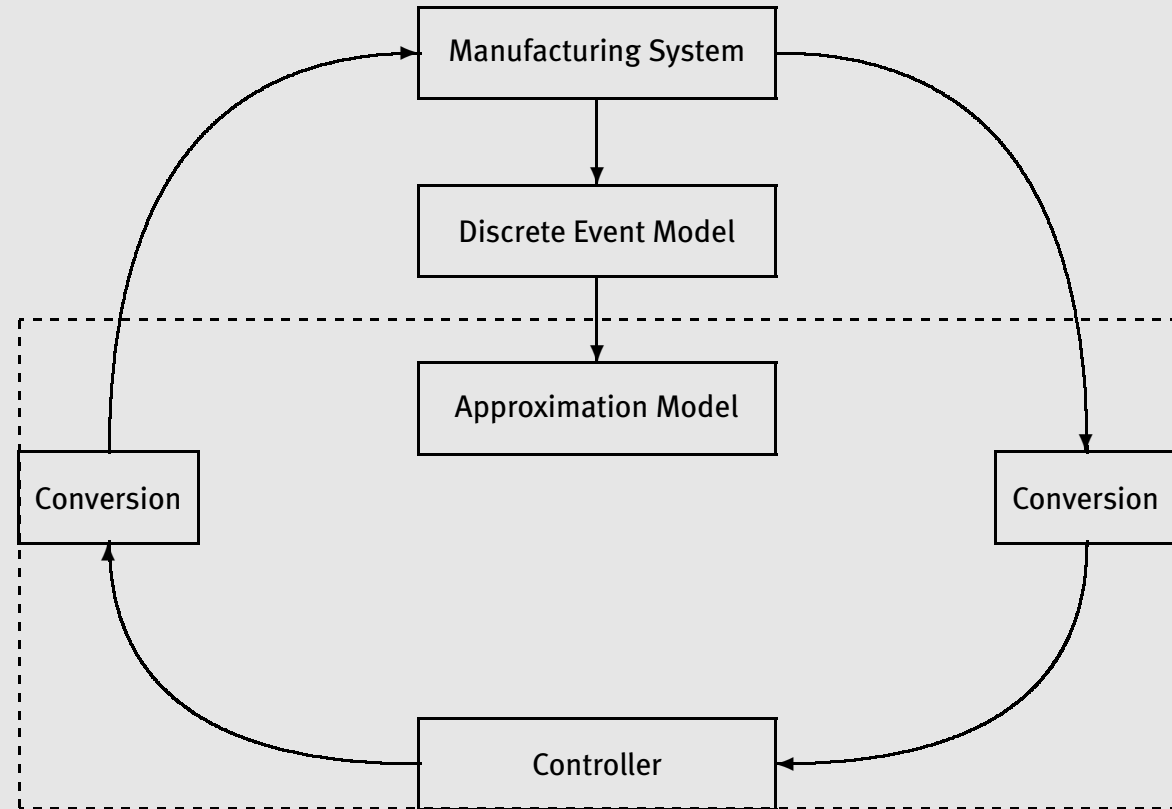
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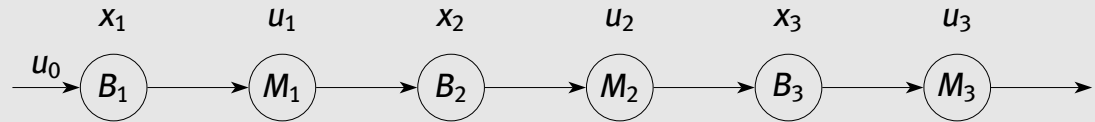
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Control Framework



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Approximation model



$$x_1(k + 1) = x_1(k) + u_0(k) - u_1(k)$$

$$x_2(k + 1) = x_2(k) + u_1(k) - u_2(k)$$

$$x_3(k + 1) = x_3(k) + u_2(k) - u_3(k)$$

or

$$\dot{x}_1(t) = u_0(t) - u_1(t) \quad \dot{x}_1(t) = u_0(t) - u_1(t)$$

$$\dot{x}_2(t) = u_1(t) - u_2(t) \quad \text{or} \quad \dot{x}_2(t) = u_1(t - \tau_1) - u_2(t)$$

$$\dot{x}_3(t) = u_2(t) - u_3(t) \quad \dot{x}_3(t) = u_2(t - \tau_2) - u_3(t)$$

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Modeling manufacturing flow

- density $\rho(x, t)$,
- speed $v(x, t)$,
- flow $u(x, t) = \rho(x, t)v(x, t)$,
- Conservation of mass: $\frac{\partial \rho}{\partial t}(x, t) + \frac{\partial \rho v}{\partial x}(x, t) = 0$.
- Boundary condition: $u(0, t) = \lambda(t)$

Modeling manufacturing flow

Armbruster, Marthaler, Ringhofer (2002):

- Single queue: $\frac{1}{v(x,t)} = \frac{1}{\mu} (1 + \int_0^1 \rho(s,t) ds)$
- Single queue: $\frac{\partial \rho v}{\partial t}(x,t) + \frac{\partial \rho v^2}{\partial x}(x,t) = 0$

$$\rho v^2(0,t) = \frac{\mu \cdot \rho v(0,t)}{1 + \int_0^1 \rho(s,t) ds}$$

- Re-entrant: $v(x,t) = v_0 \left(1 - \frac{\int_0^1 \rho(s,t) ds}{W_{\max}} \right)$

- Re-entrant: $\frac{\partial \rho v}{\partial t}(x,t) + \frac{\partial \rho v^2}{\partial x}(x,t) = 0$

$$\rho v^2(0,t) = \rho v(0,t) \cdot v_0 \left(1 - \frac{\int_0^1 \rho(s,t) ds}{W_{\max}} \right)$$

Lefeber (2003):

- Line of m identical queues: $v(x,t) = \frac{\mu}{m + \rho(x,t)}$

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Validation studies

- Line of 15 identical machines
- Infinite queues
- FIFO-policy
- Exponential Effective Processing Times
- Step-response (initially empty, start rate λ)
- Model 1, 2, 5 versus averaged discrete event

Rampup to 50% utilization (averaged discrete event)

Rampup to 50% utilization (validation studies)

From real fab to DEM

Effective process times

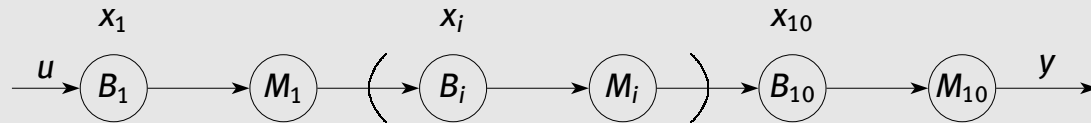
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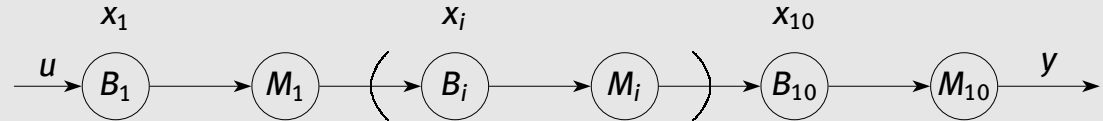
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Control



- Number of machines $n = 10$
- Infinite queues
- FIFO-policy
- Exponential Effective Processing Times
- Mean processing time: 0.5h
- Desired $u = 0.75$ (1.5 lot per h)
- Initial WIP $x_i(0) = 0$

Nonlinear model



Discretizing PDE-model yields:

$$x_1(k+1) = x_1(k) - \frac{\mu x_1(k)}{10 + x_1(k)} + u(k)$$

$$x_2(k+1) = x_2(k) - \frac{\mu x_2(k)}{10 + x_2(k)} + \frac{\mu x_1(k)}{10 + x_1(k)}$$

⋮

$$x_{10}(k+1) = x_{10}(k) - \frac{\mu x_{10}(k)}{10 + x_{10}(k)} + \frac{\mu x_9(k)}{10 + x_9(k)}$$

$$y(k) = \frac{\mu x_{10}(k)}{10 + x_{10}(k)}$$

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MPC controller design

- Prediction horizon $p = 100h$
- Control horizon $p = 5h$
- Control constant over periods of 1h
- Time sampling: 40 steps per 1h

Cost function:

$$\min_u \sum_{i=0}^p \|y(k+i|k) - y_{\text{des}}\|_Q^2$$

Constraints:

$$0 \leq u(k) \leq 2 \quad 0 \leq \frac{\mu x_i(k)}{1 + x_i(k)} \leq 2$$

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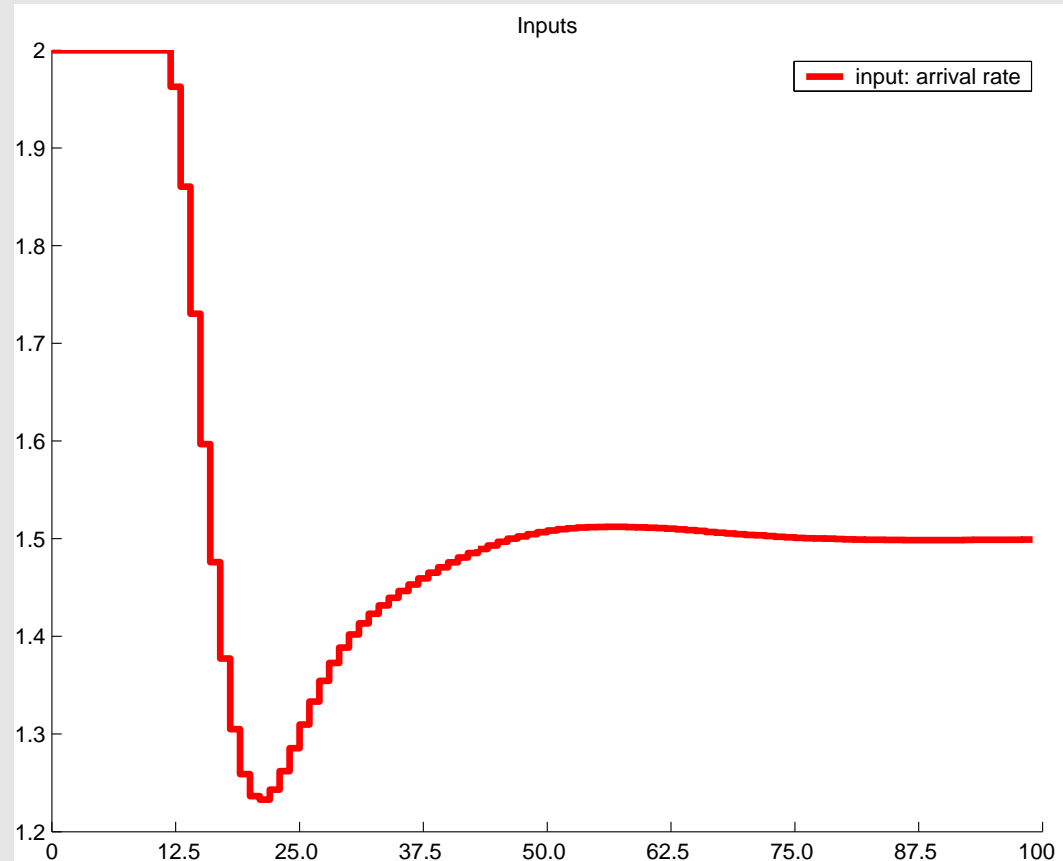
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MPC based controller design



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- EPT can be used to get from real data to simple queueing network model
- Control framework
- Control example